

1 **Three-way relationships between gut microbiota, helminth assemblages and bacterial**
2 **infections in wild rodent populations**

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29

30

Abstract

31 Background

32 Despite its central role in host fitness, the gut microbiota may differ greatly between
33 individuals. This variability is often mediated by environmental or host factors such as diet,
34 genetics, and infections. Recently, a particular attention has been given to the interactions
35 between gut bacteriota and helminths, as these latter could affect host susceptibility to other
36 infections. Further studies are still required to better understand the three-way interactions
37 between gut bacteriota, helminths and other parasites, especially because previous findings
38 have been very variable, even for comparable host-parasite systems.

39

40 Methods

41 In our study, we used the V4 region of the 16S rRNA gene to assess the variability of gut
42 bacteriota diversity and composition in wild populations of a small mammal, the bank vole
43 *Myodes glareolus*. Four sites were sampled at a regional geographical scale (100 km) along a
44 North-South transect in Eastern France. We applied analyses of community and microbial
45 ecology to evaluate the interactions between the gut bacteriota, the gastro-intestinal helminths
46 and the pathogenic bacteria detected in the spleen.

47

48 Results

49 We identified important variations of the gut bacteriota composition and diversity among
50 bank voles. They were mainly explained by sampling localities and reflected the North/South
51 sampling transect. In addition, we detected two main enterotypes, that might correspond to
52 contrasted diets. We found geographic variations of the Firmicutes/Bacteroidetes ratio, that
53 correlated positively with body mass index. We found positive correlations between the
54 specific richness of the gut bacteriota and of the helminth community, as well as between the
55 composition of these two communities, even when accounting for the influence of
56 geographical distance. The helminths *Aonchotheca murissylvatici*, *Heligmosomum mixtum*
57 and the bacteria *Bartonella* sp were the main taxa associated with the whole gut bacteriota
58 composition. Besides, changes in relative abundance of particular gut bacteriota taxa were
59 specifically associated with other helminths (*Mastophorus muris*, *Catenotaenia henttoneni*,
60 *Paranoplocephala omphalodes* and *Trichuris arvicola*) or pathogenic bacteria. Especially,
61 infections with *Neoehrlichia mikurensis*, *Orientia* sp, *Rickettsia* sp and *P. omphalodes* were
62 associated with lower relative abundance of the family Erysipelotrichaceae (Firmicutes),

63 while coinfections with higher number of bacterial infections were associated with lower
64 relative abundance of a Bacteroidales family (Bacteroidetes).

65

66 Conclusions

67 These results emphasize complex interlinkages between gut bacteriota and infections in wild
68 animal populations. They remain difficult to generalize due to the strong impact of
69 environment on these interactions, even at regional geographical scales. Abiotic features, as
70 well as small mammal community composition and within host parasite coinfections, should
71 now be considered to better understand the spatial variations observed in the relationships
72 between gut bacteriota, gastro-intestinal helminths and bacterial infections.

73

74 Keywords : bank voles, co-infections, interactions, microbial community ecology, zoonoses

75

76

Introduction

77 Vertebrate gut microbiota plays key roles in host fitness, through functions including, among
78 others, nutrient acquisition, immunity and defence against infectious exogenous agents
79 (hereafter called ‘parasites’ and including micro- and macroparasites) or proliferating
80 indigenous organisms (Belkaid & Hand, 2014; Kamada *et al.*, 2013; Round & Mazmanian,
81 2009). Nonetheless, the gut microbiota may differ greatly in natural environments between
82 individuals, populations and species (Vujkovic-Cvijin *et al.*, 2020). Its composition is even
83 subject to high temporal variation for a given individual, that is driven by stochastic processes
84 and/or variation in microbial fitness (Kolodny & Schulenburg, 2020). This variability reflects
85 hosts intrinsic factors (notably phylogeny, genetics or vertical transmission from mother to
86 offspring) as well as extrinsic features (acquisition of microorganisms from the environment,
87 potentially through diet Ley *et al.*, 2008; Moran *et al.*, 2019). It might also reveal disruption
88 of host-associated gut microbiota (termed “dysbiosis”) caused by environmental factors,
89 among which are anthropogenic pressures (e.g., chemical exposures, Rosenfeld, 2017) or
90 parasite infections (Trevelline *et al.*, 2019).

91 Understanding the relationships between gut microbiota and parasite transmission is crucial
92 regarding their potential impacts on human and animal health (Clemente *et al.*, 2012). Among
93 the numerous studies of vertebrate microbiota, some of them have put an emphasis on the gut
94 bacterial microbiota (called hereafter ‘gut bacteriota’) and their interactions with gastro-
95 intestinal helminth parasites. On one hand, the gut bacteriota may act as an innate immune
96 barrier to intestinal infections and influence the colonisation and growth of eukaryotic
97 parasites, including helminths, through competitive metabolic interactions or induction of host
98 immune responses (Leung *et al.*, 2018). On the other hand, helminth infections may also
99 affect, directly or indirectly through similar ecological and/or evolutionary processes, the
100 composition of the gut bacteriota, via physical contact, competition for resources or host
101 immunoregulation (see Kreisinger *et al.*, 2015).

102 Helminths and the gut bacteriota interactions may thus lead to positive and negative
103 interactions (Loke & Lim, 2015), with potentially local but also systemic physiological
104 changes affecting host health. For example, helminth infections can lead to malnutrition and
105 weight loss, through the dysfunction of microbial metabolism that could result from negative
106 impacts on fermentative gut bacteria (Leung *et al.*, 2018). Besides, some helminth infections
107 promote higher abundance of gut bacteria that produce short-chain fatty acids from dietary
108 fiber. These metabolites circulate throughout the body and are important regulators of host

109 physiology (glucose and fat metabolism) and immune system (Honda & Littman, 2016; Kim,
110 2021). Interactions between these helminths and gut bacteria may here increase the host anti-
111 inflammatory and regulatory T cell suppressor responses, what may in turn affect host
112 susceptibility to other infections as well as the outcomes of infections (Glendinning *et al.*,
113 2014).

114 The gut microbiota may also influence microparasite infections, through their immune
115 function against pathogenic bacteria colonization and their role in maintaining the intestinal
116 epithelium integrity (Khosravi & Mazmanian, 2013). There is also strong evidence for
117 interactions between the gut microbiota and extra-intestinal microbiota communities, at least
118 in laboratory mice. This systemic impact of gut microbiota is mediated by host immunity. As
119 such, the gut microbiota produces metabolites (e.g., bacteriocins, short-chain fatty acids,
120 microbial amino-acids...) that translocate from the intestinal lumen to various organs (e.g.,
121 liver, brain, lung) through the circulatory system. This may induce tissue-specific
122 local immune responses, and affect the host's susceptibility/resistance to (non enteric)
123 pathogens. Most of these studies have focused on viruses (e.g., influenza A, coronaviruses...)
124 and not yet on pathogenic bacteria. The systemic impact of gut bacteriota on microparasite
125 infections still represents a fundamental knowledge gap in wild animals (Pascoe *et al.*, 2017;
126 Rolhion & Chassaing, 2016).

127 The three-way interactions between host's gut bacteriota, gastro-intestinal helminths and
128 microparasites have been scarcely investigated in a single system, despite clearly becoming
129 pivotal in disease ecology. Yet, the growing interest on gut bacteriota/parasitism relationships
130 in recent literature (P. T. Johnson *et al.*, 2015) highlights the critical need for further empirical
131 works, given the relatively low concordance of findings between previous studies – even for
132 comparable host-parasite systems. Up to now, most of the research on this topic have been
133 conducted on model species in laboratory settings. Although experiments under controlled
134 conditions may help emphasizing general patterns and deciphering the mechanisms
135 underlying these interactions between gut bacteriota and parasites in vertebrates (Pascoe *et*
136 *al.*, 2017), they also have inherent limitations. On the one hand, they only included a
137 restricted number of targeted parasites (usually helminths and/or microparasites), omitting the
138 potential effects of species interactions between and within parasite communities at the intra-
139 host level (Telfer *et al.*, 2010). Co-infections by helminths species have been noticed by
140 parasitologists for decades, if not centuries. Yet, co-infections between highly divergent
141 micro- and macroparasites are also recognized to be the rule in most hosts in natural

142 environments, with wild animals that may carry simultaneously a large number of bacteria,
143 helminths, viruses (Hoarau *et al.*, 2020). On the other hand, they are intrinsically unable to
144 include – and then capture the complexity of – the environmental conditions as drivers of the
145 composition of gut bacteriota and of the exposure or sensibility to these latter (Adair &
146 Douglas, 2017). From there, studies in natural contexts deserve strong consideration because
147 these extrinsic factors may impact deeply the relationships between gut bacteriota and
148 parasitism.

149 Here, we strived to bridge these gaps by assessing the variability of gut bacteriota diversity
150 and composition in wild populations of the bank vole *Myodes glareolus*, a small mammal
151 reservoir of a large number of infectious agents (e.g., Abbate *et al.*, In_revision). We studied
152 the relationships between its gut bacteriota, parasite infracommunities – focusing on gastro-
153 intestinal helminths and pathogenic bacterial infections –, and host and environmental factors
154 that may either influence or indicate the health status (e.g., proxies such as the body mass
155 index (BMI)). The current study therefore addressed two main questions: (1) how is the
156 structure (composition and diversity) of the gut bacteriota influenced by host and
157 environmental factors? (2) does the structure of gut bacteriota also reflect interactions with
158 gastro-intestinal helminth and pathogenic bacterial communities?

159

160 **Material and methods**

161 **Data collection**

162 Bank vole sampling

163 Bank voles (*Myodes glareolus*) were trapped in summer, between late June and early
164 September 2014 in forests located in four French localities (Table 1, Figure S1) distributed
165 along a North-South transect in Eastern France, and separated by 40 to 120 km from one
166 another. The standardized trapping protocol used here was described in details in Dubois *et al.*
167 (2018).

168 Rodents were euthanized using isofluorane and cervical dislocation, as recommended by
169 Mills (1995). Morphological measures were taken. Age groups (juveniles and adults) were
170 defined according to body mass and sexual maturity. This latter was inferred using testes
171 length and position, and seminal vesicle development for males, or uterus size for females.
172 Body condition was estimated using the body mass index (BMI = weight/length²). The
173 digestive tract and the spleen were removed and stored respectively in 96% ethanol and RNA
174 later solution (-20°C).

175 Ethical statements: Animal capture and handling have been conducted according to the
176 French and European regulations on care and protection of laboratory animals (French Law
177 2001-486 issued on June 6, 2001 and Directive 2010/63/EU issued on September 22, 2010).
178 The CBGP laboratory has approval (D-34-169-003) from the Departmental Direction of
179 Population Protection (DDPP, Hérault, France), for the sampling of rodents and the storage
180 and use of their tissues.

181

182 **Table 1-** Number of bank voles analysed and prevalence of potentially pathogenic bacteria and gastro-intestinal
183 helminths for each sampling locality. N is the number of bank voles analysed (Grey cells). N_T represents the
184 number of individuals with data available for the three intra-host communities (gut bacteriota, pathogenic
185 bacteria and gastro-intestinal helminths). N_{GB} , N_{PB} and N_{GIH} respectively represent the number of individuals
186 with data available for each of these intra-host communities. ‘*Uninfected*’ corresponds to the number of
187 uninfected bank voles for a given intra-host community. ‘*Co-infection*’ corresponds to the number of bank voles
188 infected with at least two parasites of a given intra-host community. Prevalence is provided for each pathogenic
189 bacteria detected from the spleen, and each gastro-intestinal helminth. The red color gradient illustrates
190 variations in prevalence (0% = light red to 100% = dark red).
191

		Sampling localities					
		Full data	Mont-sous-Vaudrey	Chaux-des-Crotenay	Chatillon	Cormaranche	
N _T	124		22		34	36	32
Gut bacteriota							
N _G	161		37		42	41	41
Pathogenic bacteria							
N _{Pb}	138		22		34	38	37
Uninfected _{Pb}	11		3		1	3	4
Co-infection _{Pb}	67		10		21	16	20
<i>Bartonellosp</i>	0.73		0.80		0.81	0.69	0.66
<i>Mycoplasma haemomuris</i>	0.36		0.28		0.39	0.36	0.39
<i>Anaplasma phagocytophilum</i>	0.14		0.04		0.33	0.08	0.11
<i>Neoehrlichia mikurensis</i>	0.13		0.16		0.00	0.08	0.29
<i>Orientia tsutsugamushi</i>	0.08		0.00		0.08	0.21	0.00
<i>Rickettsia sp</i>	0.05		0.00		0.08	0.05	0.05
<i>Spiroplasma sp</i>	0.04		0.04		0.00	0.00	0.11
<i>Borrelia miyamotoi</i>	0.02		0.04		0.00	0.03	0.03
<i>Borrelia afzelii</i>	0.02		0.08		0.00	0.00	0.03
<i>Mycoplasma coccoides</i>	0.01		0.04		0.00	0.00	0.00
Gastro-intestinal helminths							
N _{Gih}	153		37		42	39	35
Uninfected _{Gih}	29		16		4	6	3
Co-infection _{Gih}	64		2		22	13	27
<i>Heligmosomoides glareoli</i>	0.48		0.49		0.36	0.54	0.54
<i>Heligmosomum mixtum</i>	0.41		0.00		0.76	0.00	0.86
<i>Aonchotheca murissylvatica</i>	0.35		0.03		0.26	0.59	0.54
<i>Catenotzenia henttoneni</i>	0.10		0.03		0.19	0.10	0.06
<i>Paranoplocephala omphalodes</i>	0.06		0.00		0.14	0.00	0.09
<i>Arostrilepis horrida</i>	0.03		0.00		0.00	0.00	0.14
<i>Mastophorus muris</i>	0.01		0.05		0.00	0.00	0.00
<i>Trichuris arvicola</i>	0.01		0.03		0.00	0.00	0.00

192

193

194 Characterization of gut bacteriota

195 We first characterized the gut bacteriota of bank voles. We focused on the colon as rodent gut
 196 microbiota exhibits the highest level of bacterial diversity in the lower segment of the
 197 digestive tract (Suzuki & Nachman, 2016). DNA was extracted in 2016 from a 5 mm piece of
 198 colon tissue (taken about 1 cm far from the caecum - lumen was removed) of each bank vole
 199 using the ZymoBiotics 96 DNA Kit (Zymo) following the manufacturer's instructions. We
 200 amplified a 251-bp portion of the V4 region of the 16S rRNA gene (16S-V4F

201 [GTGCCAGCMGCCGCGGTAA] and 16S-V4R [GGACTACHVGGGTWTCTAATCC]),
202 following Kozich et al. (2013) and as described in Galan et al. (2016). Samples were
203 multiplexed using dual-indexes (index i5 in the forward primer and index i7 in the reverse
204 primer). Negative controls for extraction (whole reagents without DNA), for PCR (PCR mix
205 without DNA), and for indexing (wells without reagents corresponding to particular dual-
206 indexes combinations). All DNA extractions were analysed twice using two independent
207 technical replicates of amplicon libraries. PCR products were pooled, migrated and excised on
208 a low melting agarose gel (1.25%) then purified using the NucleoSpin Gel and PCR Clean-Up
209 kit (Macherey-Nagel) and quantified using the KAPA library quantification kit (KAPA
210 Biosystems) standardized to 4nM by qPCR spectrophotometry (assay). Sequencing was
211 performed on a 251-bp paired-end Illumina MiSeq run. The raw sequence reads (.fastq
212 format) have been deposited in the Zenodo Repository.
213 Sequence data were processed as described in Galan et al. (2016) using the pipelines
214 implemented in FROGS (Find Rapidly OTU with Galaxy Solution, Escudié *et al.*, 2018).
215 Briefly, the paired-end sequences were trimmed with CUTADAPT (Martin, 2011), merged with
216 FLASH (MAGOC & SALZBERG, 2011), and clustered into fine-scale molecular operational
217 taxonomy OTU units at 97% identity using the SWARM algorithm (Mahe *et al.*, 2014)
218 executed with aggregation parameter distance $d=1$ and a second pass performed on the seeds
219 of previous clusters with $d=3$. As such, OTUs do not correspond to a fixed clustering
220 threshold but rather to clusters that are close to amplicon sequencing variants (ASVs).
221 Putative chimeras were removed using VSEARCH tools with de novo VUCHIME and the cross-
222 validation method. Taxonomy was assigned with BLASTN+ (Camacho *et al.*, 2009) using the
223 SILVA SSU Ref NR 128 database as a reference (<http://www.arb175silva.de/projects/ssu-ref-nr/>). Filtering for false positives was carried out as proposed by Galan et al. (2016). In short,
225 we discarded positive results associated with sequence counts below two OTU-specific
226 thresholds, which checked respectively for cross-contamination between samples (using the
227 negative controls for extraction and PCR) and incorrect assignment due to the generation of
228 mixed clusters on the flow cell during Illumina sequencing, using a false index-pairing rate
229 for each PCR product of 0.02%, based on estimates from Galan et al. (2016). For each
230 sample, only OTUs found in the two technical replicates were considered as positive, and
231 OTUs found in only one of the two replicates were removed. The number of sequences
232 obtained for each technical replicate from a sample were summed.

233 Lastly, we discarded OTUs and samples containing less than 500 reads in the dataset, as well
234 as OTUs considered to be contaminants, following (Salter *et al.*, 2014). OTUs number of
235 reads were finally normalised to proportional abundance within each rodent (McKnight *et al.*,
236 2019). We only considered the family taxonomic rank for further analyses, but analyses at the
237 phylum level provided similar results (not shown).

238

239 Detection of pathogenic bacteria and gastro-intestinal helminths
240 We described the presence/absence of pathogenic bacteria from the spleen of each bank vole.
241 This lymphoid organ filters microbial cells in mammals and as such, enables to recover recent
242 infections (Abbate *et al.*, In_revision; Diagne *et al.*, 2017). Molecular protocols,
243 bioinformatics pipelines and data filtering were similar to those described above (gut
244 bacteriota), except for the DNA extraction from splenic tissue using DNeasy 96 Tissue Kit
245 (Qiagen). The potential pathogenicity of each bacterial OTU was assessed based on published
246 literature and on the Gideon database (<https://www.gideononline.com/>). Opportunistic
247 pathogens (*i.e.* commensal agents in healthy hosts, that become pathogenic when the balance
248 of the immune system is disrupted) were discarded from the dataset. Only the information of
249 the presence / absence of pathogenic OTUs was considered. For each bank vole, helminths
250 were carefully extracted and counted from the different sections of the digestive tract
251 (stomach, small intestine, large intestine and caecum), and classified by morphotype then
252 stored in 95% ethanol for further accurate identification. The latter was based on
253 unambiguous morphological criteria using conventional microscopy and generalist
254 identification keys or specific literature when available (Anderson *et al.*, 2009 ; Khalil *et al.*,
255 1994 ; Ribas Salvador *et al.*, 2011).

256

257 **Statistical analyses**

258 All statistical analyses were implemented in R v4.0.3 (team, 2020). For more convenience,
259 gut bacteriota, pathogenic bacteria and gastro-intestinal helminths were further described as
260 ‘intra-host communities’.

261

262 Gut microbiota diversity and composition

263 Description and analyses of bacterial communities were performed using the PHYLOSEQ
264 package (McMurdie & Holmes, 2013). We considered three features to analyse within hosts’
265 gut microbiota. i) We looked for enterotypes, *i.e.* distinct community composition types of gut

266 bacteriota, as found in humans (Arumugam *et al.*, 2011; Holmes *et al.*, 2012), using the
267 Dirichlet Multinomial Mixtures DMM (Morgan, 2021). ii) We analysed the Firmicutes
268 /Bacteroidetes (F/B) log-ratio, as it is often used as a proxy of health or metabolism in humans
269 and mice, see refs in (Lavrinienko *et al.*, 2018). We calculated this ratio with the MICROBIOTA
270 package (Lahti & Shetty, 2017). iii) We characterized the alpha diversity using two metrics,
271 the specific richness (*i.e.* number of taxa within the host individual) and the Shannon index as
272 recommended in (Haegeman *et al.*, 2013).

273 We estimated the beta diversity, *i.e.*, the dissimilarity between host individuals in their gut
274 bacteriota using Bray-Curtis distances. We considered the relative abundance of OTUs
275 (family).

276

277 Influence of host and environmental factors on gut bacteriota diversity and composition
278 We tested the influence of individual characteristics (age class, gender, BMI) and localities,
279 independently on the F/B log-ratio and on the alpha diversity using generalized linear models
280 (GLM). We considered a negative binomial error distribution for the F/B ratio and the specific
281 richness, and a gaussian distribution for the Shannon index. Best model selection was
282 performed considering models with all possible combinations of factors and the DREDGE
283 function of the MUMIN package. The best model was selected using the Akaike information
284 criterion corrected for small sample size AICc, (J. B. Johnson & Omland, 2004). We assessed
285 the effect of each factor in the best model with the Δ AICc index. When the factor locality was
286 significant, Tukey's post-hoc tests were applied to evaluate pairwise differences between
287 localities, using the MULTCOMP package (Hothorn *et al.*, 2008). Residuals were checked to
288 graphically to ensure that all assumptions regarding normality, independence and the
289 homogeneity of variance were satisfied.

290 We evaluated the influence of geographic distance on the dissimilarities in gut bacteriota by
291 performing Mantel tests and using Pearson correlation (10,000 permutations). These tests
292 have less statistical power to address questions related to the variation in community
293 composition data among sites. Therefore, we also analysed the factors shaping the
294 dissimilarities in gut microbiota composition using several functions of the VEGAN package
295 (Oksanen *et al.*, 2020). Distance-based redundancy analyses (db-RDA) were performed to
296 analyse the effect of individual explanatory factors (age class, gender, BMI) and sampling
297 localities on dissimilarities in gut microbiota composition. Redundancy analyses are
298 appropriate to test hypotheses about the origin and maintenance of the variation in β diversity

299 (Legendre et al. 2005). We used the CAPSCALE function, followed by permutational
300 multivariate analyses of variance (PERMANOVA). We selected the best model, *i.e.*, the most
301 parsimonious one, using the ORDIR2STEP function (*P*-value adjusted and R^2 adjusted). For
302 each factor, we evaluated the intra-group dispersion using the BETADISPER function as
303 PERMANOVA analyses are sensitive to differences in dispersion among groups. A Tukey's
304 test was done to see if and which groups differed in relation to their variances. Lastly, we used
305 DESEQ2 package (Love et al., 2014) to identify the changes in bacteria taxa that best
306 explained gut bacteriome dissimilarities between individuals and localities. We performed
307 GLMs with negative binomial error (NBINOMWALDTEST method) and significant differences
308 were obtained after Benjamini & Hochberg corrections. They were visualised using the
309 METACODER package (Foster et al., 2017).

310

311 Relationships between gut bacteriota and pathogenic communities
312 We estimated the alpha diversity of the gastro-intestinal helminth and pathogenic bacteria
313 community using the richness index (presence/absence data). We used GLMS and model
314 selection process described above to analyse whether the alpha diversity of each intra-host
315 community (gut bacteriota and pathogenic communities) was influenced by the alpha
316 diversity of the two other ones.

317 We estimated the beta diversity of the gastro-intestinal helminth and pathogenic bacteria
318 community using the Jaccard index (presence/absence data). The relationships between intra-
319 host community dissimilarities were investigated using three approaches. i) We applied
320 partial Mantel tests using MULTI.MANTEL (phytools package Revell, 2012) to analyse the
321 correlation between two matrices of dissimilarities (corresponding to two different
322 communities), while controlling for the effect of a third dissimilarity matrix (third
323 community). ii) We used db-RDA to analyse more deeply the relationships between the gut
324 bacteriota and the pathogenic (bacteria and helminths) communities. We included the alpha
325 diversity indices (richness specific) and infectious status as presence / absence) of pathogens
326 with prevalence greater than 10% in at least one locality as explanatory variables in these
327 analyses. We selected the best model using the ORDIR2STEP method. iii) We used DESEQ2 to
328 determine the gut bacteria taxa whose relative abundances changed with significant
329 explanatory variables.

330

331

332

Results

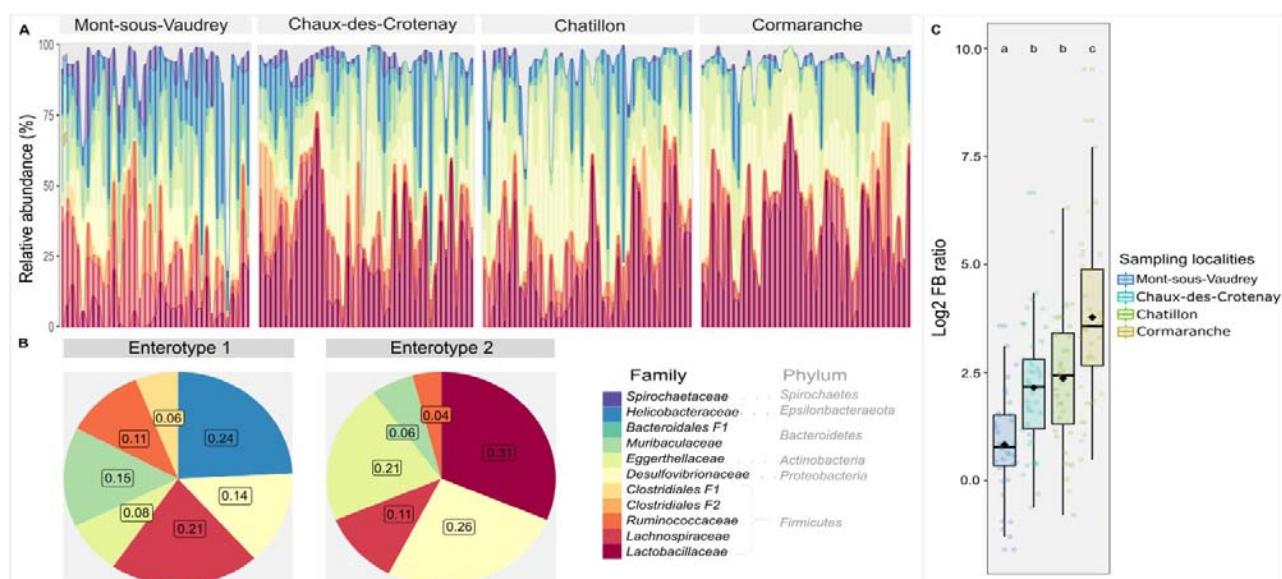
333

334 A total of 186 bank voles were trapped during the fieldwork campaign over the four targeted
335 localities. For technical reasons (e.g., poor sample preservation, missing data), we could study
336 the three intra-host communities for 124 rodents only.

337

338 Characterization of the gut bacteriota: taxa and enterotypes

339 Once the quality control steps were applied, the gut bacteriota dataset included 161 bank
340 voles. We detected 10 phyla and 61 families of bacteria. At the phylum level, we found six
341 predominant taxa that represented 99% of the gut bacteria relative abundance (Figure S2). At
342 the family level, 11 families represented 93% of the relative abundance of the gut bacteriota
343 (Figure 1A).



344

345 **Figure 1-** Composition of the intestinal bacteriota. A) The bar plot shows the individual variations of 11
346 bacterial families (F= Unknown family) belonging to 6 phyla and representing 93% of the total composition.
347 Individuals (bars) are grouped by sampling localities, which are ordered from North to South. Each color
348 represents a taxa. B) The composition of the two enterotypes identified using Dirichlet multinomial mixtures
349 (DMMs), at family rank, is shown. Bacterial families are represented using the same colors as in A. C) the ratio
350 (Firmicutes / Bacteroidetes) is shown for each sampling locality. Box and whisker plots represent the median and
351 interquartile values. Black dots correspond to the mean value, and colored dots correspond to individuals.
352 Different letters indicate statistically significant differences at $P < 0.05$, with pairwise Tukey post hoc
353 adjustments.

354

355 We distinguished two enterotypes from the DMM approach. One (enterotype 1) was mainly
356 composed of the families *Helicobacteraceae*, *Lachnospiraceae*, *Muribaculaceae* and
357 *Desulfovibrionaceae*, while the other (enterotype 2) mainly included *Lactobacillaceae*,
358 *Desulfovibrionaceae* and *Eggerthellaceae* (Figure 1B).

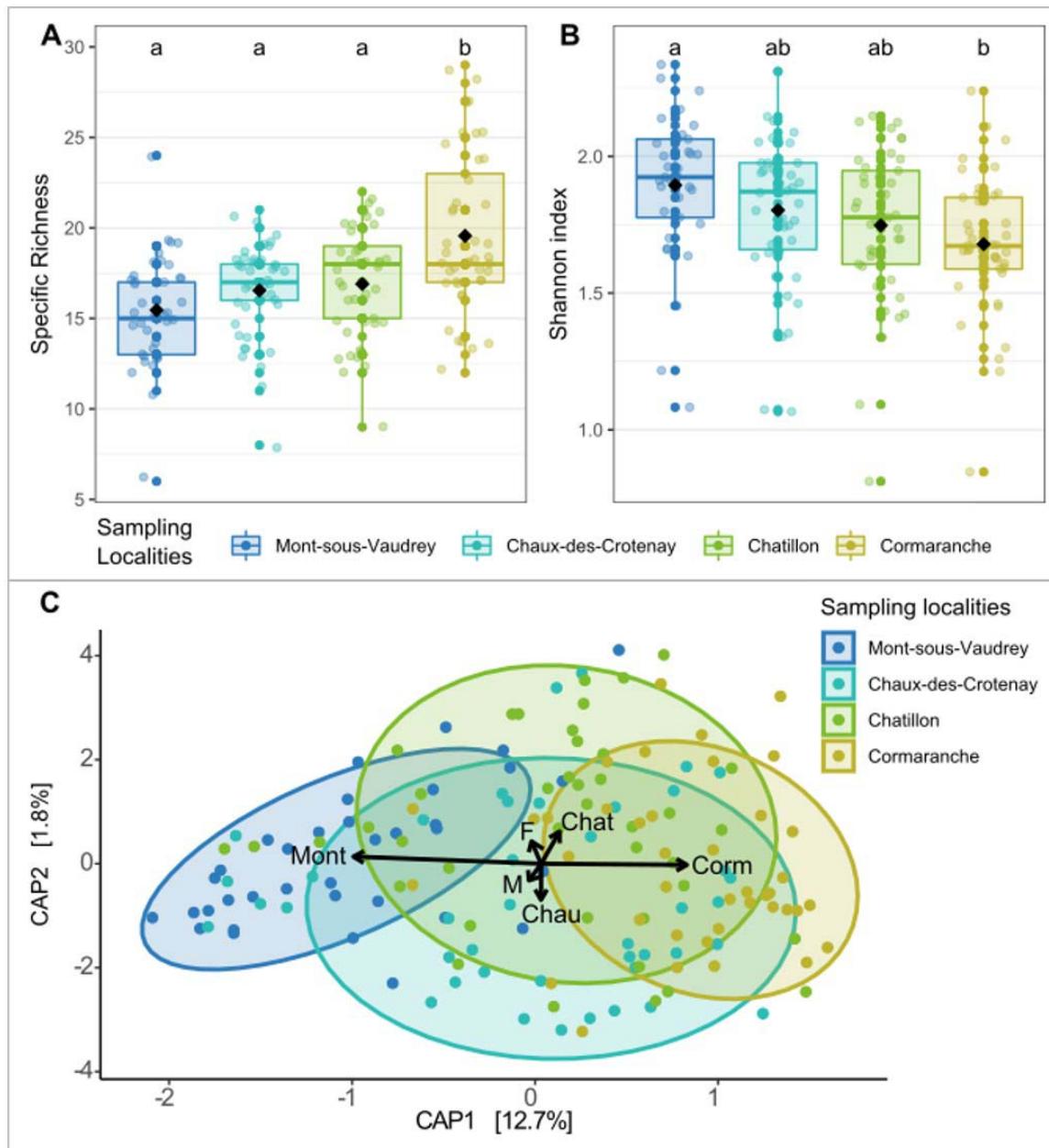
359

360 **Diversity of the gut bacteriota : the influence of sampling locality and host condition**

361 We found that the *Firmicutes* / *Bacteroidetes* ratio varied significantly between localities
362 (Figure 1C). Overall, northern localities exhibited lower F/B ratio than southern ones, with all
363 pairwise comparisons being significant except Chatillon *versus* Chaux-des-Crotenay.

364 Individual characteristics did not influence this ratio (Table S1).

365 The sampling locality had a significant global effect on the alpha diversity of the gut
366 bacteriota (GLMs. Specific richness: $F = 8.49$, $P < 10^{-3}$; Figure 2A; Shannon index: $F =$
367 4.74 , $P = 3 \times 10^{-3}$; Figure 2B; Table S2A). The locality Cormaranche exhibited a higher
368 specific richness than all other localities (*Tukey post hoc test*. Mont-sous-Vaudrey: $Z = 5.13$,
369 $P_{adj} < 10^{-3}$, Chaux-des-Crotenay: $Z = 4.57$, $P_{adj} < 10^{-3}$ and Chatillon: $Z = 3.62$, $P_{adj} = 1.7 \times 10^{-3}$)
370 but a lower level of diversity than Mont-sous-Vaudrey when considering taxa relative
371 abundance (*Tukey post hoc test*. Shannon index: $Z = -3.64$, $P_{adj} = 1.5 \times 10^{-3}$, Figure 2B). Body
372 condition (BMI) was also found to have a significant effect, but only when considering
373 specific richness ($t = 2.91$; $P = 4 \times 10^{-3}$) – with higher values of BMI associated with
374 increasing species richness. All these results are detailed in Table S2A and Figures S3).



375

376 **Figure 2-** Variations of the gut bacteriota alpha diversity between localities. Alpha diversity is represented using
 377 A) the specific richness of the gut bacteriota, and B) the Shannon index of the gut bacteriota, Results are shown
 378 per locality, ordered from North to South. Each colored point represents an individual. Black points indicate the
 379 average alpha diversity per locality. Box-and-whisker plots represent the median and interquartile values.
 380 Different letters denote statistically significant differences at $P < 0.05$, with pairwise post-hoc Tukey
 381 adjustments. C) Distance-based redundancy analysis (db-RDA) of the gut bacteriota (family) based on Bray-
 382 Curtis dissimilarities. Dots represent individuals. The colors and shapes of the dots are associated with different
 383 factors : Localities from North to South (Mont-sous-Vaudrey: Mont, Chaux-des-Crotenay: Chau, Chatillon:
 384 Chat and Cormaranche: Corm) and gender (females: F and males: M). Significatif factors based on the

385 ordiR2step analysis are shown as arrows. Ellipses represent a 80% confidence interval around the centroid of the
386 clusters, for each locality.

387

388 **Composition of the gut bacteriota : sampling locality as the main factor of variation**

389 We found a significant positive relationship between the dissimilarities in gut bacteriota
390 composition and the geographic distance (*Mantel test*. $r = 0.25$, $P = 10^{-4}$, Table S3A).
391 We found a significant effect of the sampling localities (*db-RDA*. $R^2_{adj} = 0.16$, $P = 1 \times 10^{-3}$)
392 and host gender (*db-RDA*. $R^2_{adj} = 0.01$, $P = 0.027$) on gut microbiota composition. The CAP1
393 axis discriminated Mont-sous-Vaudrey and Cormaranche localities (12.7% of the total
394 variance, Figure 2C). However, this result has to be taken cautiously as significant differences
395 of data dispersion were detected between localities (*betadisper*. $P = 1 \times 10^{-3}$). The locality
396 Cormaranche showed a lower dispersion compared to all other localities (Tukey multiple
397 comparisons, Table S3B).

398 We detected significant differences in the relative abundance of specific taxa using DeSeq2
399 (Table 2; Table S3C). The main changes (Log2 fold values higher than 20) were detected
400 between the northern (Mont-sous-Vaudrey) and southern localities. The northern population
401 was involved in 75% of all significant pairwise differences (Log2 fold change in composition
402 > 10). The gut bacteriota of these bank voles includes less Clostridiales (one unknown family;
403 Firmicutes), Bifidobacteriaceae (Actinobacteria) and Desulfovibrionales (two unknown
404 families; Proteobacteria), but more Erysipelotrichaceae (Firmicutes) than in all the three other
405 southern populations. The gut bacteriota of bank voles from Cormaranche (South) is
406 characterized by less Erysipelotrichaceae (Firmicutes), than in the three northern populations.

407

408 **Table 2**- Pairwise comparisons of the relative abundance of the gut bacteriota between sampling localities. Mont
409 = Mont-sous-Vaudrey, Chau = Chaux-des-Croteay, Chat = Chatillon, Corm = Cormaranche. The Log² fold value
410 is indicated for significant changes in abundance between two localities. Blue and red colors respectively
411 correspond to negative and positive values. Higher absolute changes in Log² fold are emphasized with darker
412 colors. The notation "order_fx" or "class_fx" is used when there was no assignation at the family level with the
413 SILVA database. Phylum is indicated in bold.

414

Taxonomy	Mont vs Chau	Mont vs Chat	Mont vs Corm	Chau vs Chat	Chau vs Corm	Chat vs Corm
<i>Epsilonbacteraeota</i>						
<i>Helicobacteraceae</i>	0	0	1.91	0	0	0
<i>Proteobacteria</i>						
<i>Desulfovibrionaceae</i>	-1.3	-1.63	-1.54	0	0	0
<i>Firmicutes</i>						
<i>Ruminococcaceae</i>	0	0	1.48	0	1.5	1.08
<i>Lactobacillaceae</i>	0	0	-1.81	0	0	0
<i>Clostridiales</i>	0	0	2.16	0	2.27	1.43
<i>Clostridiales_F2</i>	-26.15	-26.31	-26.54	0	0	0
<i>Spirochaetes</i>						
<i>Spirochaetaceae</i>	0	0	2.39	0	0	0
<i>Actinobacteria</i>						
<i>Eggerthellaceae</i>	0	-1.31	-2.42	0	-1.74	0
<i>Bacteroidetes</i>						
<i>Bacteroidales_F4</i>	0	0	0	0	0	-3.59
<i>Muribaculaceae</i>	0	0	1.74	0	1.93	1.64

415
416

417 Differences in the composition of the gut bacteriota between males and females bank voles
418 were driven by the phylum *Firmicutes*, with males exhibiting higher relative abundance of
419 this taxa than females (Table S3C).

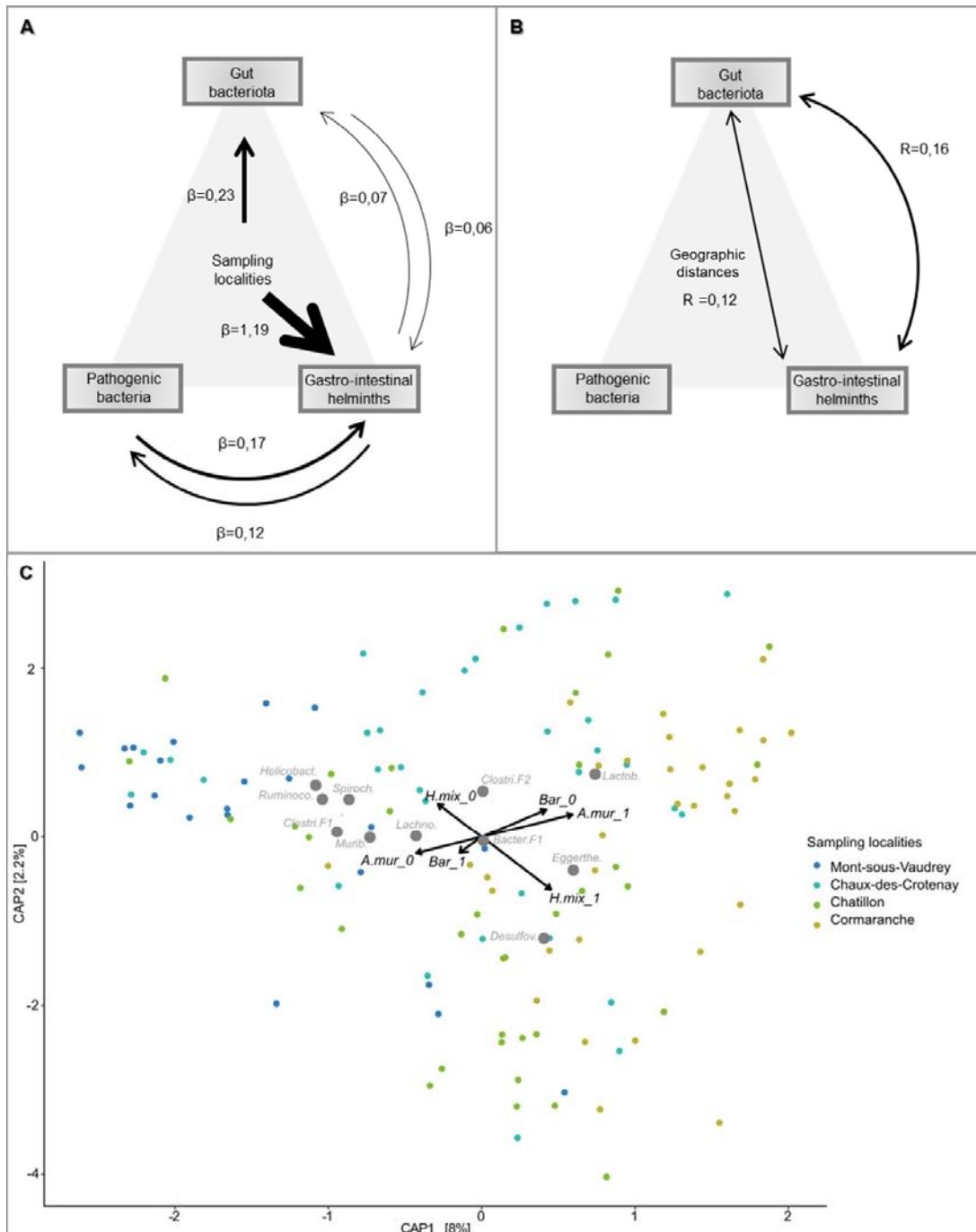
420

421 Relationships between the diversity of the three intra-host communities

422 We found a significant relationship between the specific richness of the gut bacteriota and the
423 richness of the helminth community. A more diverse gut bacteriota was associated with a
424 greater number of helminth species infecting bank voles (*GLM*. $F = 14.09$, $P < 10^{-3}$; Figure
425 3A; Table S2B).

426 We also found a positive relationship between the specific richness of the pathogenic bacteria
427 and the richness of the gastro-intestinal helminth community (*GLM*. $F = 6.99$, $P = 9 \times 10^{-3}$;
428 Figure 3A; Table S2B).

429 Lastly, we found a significant effect of the specific richness of both the gut bacteriota and of
430 pathogenic bacteria on the richness of the gastro-intestinal helminth community (*GLM*. Gut
431 bacteriota. $t = 3.50$, $P < 10^{-3}$; pathogenic bacteria $t = 2.38$, $P = 0.019$; Figure 3A, Table S2B).



432
433
434 **Figure 3- Associations between the diversity and composition of the three intra-host communities.** The two
435 upper diagrams show the relationships between A) the specific richness and B) the composition of the three
436 communities, while taking into account the influence of the sampling localities or distance geographic. The
437 effect size and the direction of the relationship between communities are represented using an arrow, its width
438 corresponding to the estimate (β) $\times 10$ or the correlation indice $R \times 10$. Only significant effects are represented.
439 C) This db-RDA triplot shows the structure of the gut bacteriota and the correlations with the pathogen
440 communities. The arrows correspond to the significant explanatory variables. Each point corresponds to an
441 individual, and the colors correspond to the different sampling localities. A.mur: *Aonchotheca murissylvatici*,
442 Bar: *Bartonella* sp, H.mix: *Heligmosomum mixtum*; Helicobacter.: Helicobacteraceae,
443 Spiroch.: Spirochaetaceae, Clostri.F2: Clostridiales_f2, Clostri.F1: Clostridiales_f1, Lactob.: Lactobacillaceae,

444 Eggerthe.: Eggerthellaceae, Desulfov.: Desulfovibrionaceae , Bacter.F1: Bacteroidales_f1,
445 Lachno.:Lachnospiraceae; Murib.: Muribaculaceae, Ruminoco.: Ruminococcaceae
446

447 **Relationships between the composition of the three intra-host communities**

448 We found a positive relationship between dissimilarities in the gut bacteriota and
449 dissimilarities in the gastro-intestinal helminth community composition (*partial Mantel test. r*
450 $= 0.16, P = 1 \times 10^{-4}$; Figure 3B), but not with dissimilarities in the pathogenic bacteria
451 community composition (*partial Mantel test. r = 0.02, P = 0.28*). After controlling for
452 geographic distances, dissimilarities in gut bacteriota composition remained significantly
453 correlated with dissimilarities in helminth community composition (*partial Mantel test. r*
454 $= 0.12; P = 0.001$). Further details are provided in Table S3A.

455 We detected significant associations between the whole composition of the gut bacteriota and
456 the presence / absence of three pathogens: *Aonchotheca murissylvatici* (*ordistep db-RDA*.
457 $R^2_{adj} = 0.0, P = 2 \times 10^{-3}$), *Bartonella* sp (*ordistep db-RDA*. $R^2_{adj} = 0.01, P = 0.04$) and
458 *Heligmosomum mixtum* (*ordistep db-RDA*. $R^2_{adj} = 0.06, P = 2 \times 10^{-3}$). The db-RDA triplot
459 based on the two first axes only represented 10.2% of the total variance (Figure 3C; Figure
460 S4). It showed that individuals infected with *Aonchotheca murissylvatici*
461 or *Heligmosomum mixtum*, but not infected with *Bartonella* sp., had
462 more *Lactobacillaceae* (*Firmicutes*), *Desulfovibrionaceae* (*Proteobacteria*)
463 and *Eggerthellaceae* (*Actinobacteria*). These individuals also had less *Spirochaetaceae*
464 (*Spirochaeta*), *Muribaculaceae* (*Bacteroidetes*), *Helicobacteraceae* (*Epsilonbacteraeota*)
465 and *Ruminococcaceae* (*Firmicutes*). This pattern is correlated with the sampling localities.
466 Individuals from northern localities are distributed on the left side of the CAP1 axis, and
467 southern ones on the right side of it (Figure 3C). Neither the specific richness of pathogenic
468 bacteria nor the specific richness of the gastro-intestinal helminth community had a
469 significant effect on the global composition of the gut bacteriota (Table S3D).

470
471 The specific richness of the gastro-intestinal helminth community, as well as infections with
472 *A. murissylvatici* and *H. mixtum*, were only slightly associated with different relative
473 abundance of particular gut bacteria taxa (*DeSeq2*. Log2 fold changes did not exceed 3.5).
474 These changes concerned four main families. Rhizobiaceae and Spirochaetaceae showed
475 negative associations with gastro-helminth specific richness and *A. murissylvatici*. Mollicutes
476 (undetermined family) and Saccharimonadaceae showed positive associations with *A.*
477 *murissylvatici* and *H. mixtum* (Table 3). More details are provided in Table S3E.
478

479 **Table 3-** Changes in relative abundance of the gut bacteriota with regard to infectious status (helminths and
480 pathogenic bacteria) and specific richness. The Log² fold change in relative abundance is indicated for
481 significant values only. Negative values are represented with blue colors, positive values with red colors.
482 Higher absolute changes in Log² fold are emphasized with darker colors.

Phylum	Family	log2FoldChange
Richness of GI helminths		
Actinobacteria	Eggerthellaceae	0.67
Proteobacteria	Rhizobiaceae	-3.79
Epsilonbacteretaeota	Helicobacteraceae	-0.89
<i>Aonchotheca murissylvatici</i>		
Proteobacteria	Desulfovibrionaceae	1.13
Tenericutes	Mollicutes_f1	3.07
Patescibacteria	Saccharimonadaceae	3.03
Spirochaetes	Spirochaetaceae	-1.75
Actinobacteria	Eggerthellaceae	1.4
<i>Heligmosomum mixtum</i>		
Tenericutes	Mollicutes1	3.34
<i>Mastophorus muris</i>		
Proteobacteria	Desulfovibrionales3	-29.98
Firmicutes	Clostridiales_f1	-21.41
Firmicutes	Clostridiales_f2	-23.73
<i>Catenotaenia henttoneni</i>		
Proteobacteria	Rhizobiaceae	-24.71
Proteobacteria	Burkholderiaceae	-23.96
Firmicutes	Erysipelotrichaceae	-20.6
<i>Paranoplocephala omphalodes</i>		
Firmicutes	Clostridiales_f2	-25.37
Proteobacteria	Desulfovibrionales_f3	-23.74
Proteobacteria	Rhizobiaceae	-24.86
Proteobacteria	Burkholderiaceae	-22.69
Firmicutes	Erysipelotrichaceae	-22.4
Firmicutes	Christensenellaceae	-5.24
<i>Trichuris arvicola</i>		
Firmicutes	Christensenellaceae	-20.69
Firmicutes	Clostridiales_f1	-18.62
Bacteroidetes	Bacteroidales_f4	-18.38
Richness of Pathogenic bacteria		
Bacteroidetes	Bacteroidales_f3	-28.51
<i>Neoehrlichia mikurensis</i>		
Firmicutes	Erysipelotrichaceae	-23.42
Firmicutes	Clostridiales	-2.72
Bacteroidetes	Muribaculaceae	-1.76
<i>Orientia tsutsugamushi</i>		
Firmicutes	Erysipelotrichaceae	-22.95
Proteobacteria	Desulfomicrobiaceae	-5.71
Bacteroidetes	Muribaculaceae	-2.45
<i>Anaplasma phagocytophilum</i>		
Firmicutes	Ruminococcaceae	-1.75
<i>Rickettsia sp</i>		
Firmicutes	Erysipelotrichaceae	-21.81

.1

484

485 In the opposite, we found strong changes in relative abundance of gut bacteriota families with
486 other gastro-intestinal helminths and pathogenic bacteria infections (*DeSeq2*. Log2 fold
487 higher than 18, Table 3). It concerned infections with the helminth species *Mastophorus*
488 *muris*, *Catenotaenia henttoneni*, *Paranoplocephala omphalodes* and *Trichuris arvicola*.
489 These associations were all negative and mostly involved the same bacterial families (Table
490 S3E), namely undetermined families of Bacteroidales (Bacteroidetes), Desulfovibrionales
491 (Proteobacteria) or Clostridiales (Firmicutes), Erysipelotrichaceae (Firmicutes), Rhizobiaceae
492 (Proteobacteria) and Burkholderiaceae (Proteobacteria).

493 Considering pathogenic bacteria, we found that higher levels of specific richness were
494 associated with lower relative abundance of an undetermined Bacteroidales family
495 (Bacteroidetes), and that *Neohhrlichia mikurensis*, *Orientia tsutsugamushi* and *Rickettsia* sp
496 infections were associated with strong decreases in relative abundance of *Erysipelotrichaceae*
497 (Firmicutes) (Table 3; Table S3E). Other associations between bacterial infections and
498 changes in relative abundance of specific gut bacteriota taxa were detected, but with little size
499 effect (*DeSeq2*. log2 fold changes lower than 5).

500

501 Discussion

502 Understanding the complex interlinkages between host microbiota, host-pathogen interactions
503 and health in wild animal populations has become a key topic in disease ecology, with
504 consequences for population dynamics, zoonotic risk management or biodiversity
505 conservation. Here, we use a combination of metabarcoding and community ecology
506 approaches to describe the gut microbiota of wild rodent populations and their variations at a
507 regional geographical scale, and to explore the three-way relationships between the gut
508 bacteriota and communities of gastro-intestinal helminths and pathogenic bacteria.

509

510 Spatial variations of gut bacteriota and their potential causes

511 The gut microbiota of bank voles has been mainly examined in the context of exposure to
512 radioactive pollutants (e.g., Lavrinienko (2018), but see Knowles et al. (2019)). In this study,
513 we focused on localities sampled at a regional scale (100 km) along a North-South gradient in
514 Eastern France.

515 We found significant inter-individual variations in the gut bacteriota composition although
516 intrinsic factors such as gender and age played little role. Interestingly, we found that all

517 individuals were clustered within two distinct enterotypes (Arumugam *et al.*, 2011).
518 Enterotypes have already been described in wild rodents (Goertz *et al.*, 2019; Li *et al.*, 2016),
519 and they might reflect distinct ways of generating energy from substrates available in the
520 digestive tract, as well as differences in diet (Rinninella *et al.*, 2019; Wang *et al.*, 2014). In
521 bank voles, these enterotypes could be associated with different diets, one oriented toward
522 seeds and plants and another one toward insects and berries. Indeed, enterotype 1 is
523 characterized by families (namely, Helicobacteraceae, Lachnospiraceae and Muribaculaceae)
524 that are involved in the breakdown of carbohydrates, fermentation of plant saccharid and
525 degradation of glycan (see refs in Goertz *et al.*, 2019). These families are also predictive
526 signals of a high-fat diet in mice (Bowerman *et al.*, 2021; Rodriguez-Daza *et al.*, 2020). In the
527 opposite, enterotype 2 is characterized by families (namely, Lactobacillae and
528 Eggerthellaceae) that can be involved in the digestion of fermented food (e.g., rodent food
529 store over winter), and insect skeleton (see refs in Maurice *et al.*, 2015) or in the degradation
530 of polyphenol (Rodriguez-Daza *et al.*, 2020). All these aliments have varying nutritional and
531 chemical composition and may be part of bank vole diet (Ecke *et al.*, 2018). The fraction of
532 these different types of resources in bank vole diet may vary with resource preference or
533 availabilities, reproductive status, sampling date and location (e.g., Maurice *et al.*, 2015). It
534 would be interesting to develop semi-natural experiments to survey rodent diet and gut
535 microbiome through time, and analyse the link with enterotypes in bank voles (Wang *et al.*,
536 2014).
537 There are now many evidence that geographical location is likely to shape variations in gut
538 bacteriota composition between localities sampled and studies. Previous works have already
539 shown that the structure of rodent gut microbiota varied between localities at large spatial
540 scales due to biogeographic or genetic factors (Linnenbrink *et al.*, 2013). Geographic
541 variability has also been found at smaller spatial scales (e.g., few km Goertz *et al.*, 2019).
542 Here, our results provide significant evidence for spatial structure of gut bacteriota between
543 bank vole populations that are between 50 and 130 km away, with no clear barrier to dispersal
544 or gene flow (Dubois *et al.*, 2018).
545 We observe gradual changes in terms of gut bacteriota richness, evenness, composition and in
546 particular Firmicutes/Bacteroidetes ratio, between bank voles from the northern and southern
547 populations. Although the links between the diversity and functional capacity of the gut
548 bacteriota are still not fully understood (Worsley *et al.*, 2021), it is largely assumed that
549 changes in diversity are associated with shifts in metabolism (Reese & Dunn, 2018). Bank

550 voles from southern populations exhibit higher specific richness and lower evenness of the gut
551 bacteriota, as well as lower dispersion of gut bacteriota composition. They have higher levels
552 of body condition and F/B ratio, that are indicative of an optimisation of calorie intake and
553 absorption, weight gain and fat storage (see refs in Wolf *et al.*, 2021). Altogether, these results
554 could suggest strong constraints on gut bacteriota function to maximise energy extraction. The
555 northern populations show the opposite patterns. Lower BMI and lower levels of F/B ratio
556 might reflect energy production and conversion, amino acid transport and metabolism, while
557 diversity patterns (higher evenness and lower specific richness of the gut bacteriota) could
558 suggest lower stochasticity and/or directional selection. Further studies are required to
559 investigate the ecological processes driving these changes in gut bacteriota.

560 Lastly, these differences in gut microbiota composition between the northern and southern
561 populations might also reflect physiological variations related to physiology, health, and
562 potentially to immunity. Clostridiales and Bifidobacteriaceae participate in the maintenance of
563 intestinal homeostasis, and in the regulation of inflammation or in the gut barrier function
564 (Arboleya *et al.*, 2016; Hakansson & Molin, 2011; Lopetuso *et al.*, 2013), while specific taxa
565 within Erysipelotrichaceae may be correlated with inflammation or have immunogenic
566 potential (Kaakoush, 2015; Zhai *et al.*, 2019). Desulfovibrionales activities result in the
567 production of H₂S, that in turn, leads to damages of the gut barrier, production of endotoxins
568 and pro-inflammatory cytokines (Hu *et al.*, 2022). Our previous work revealed that bank voles
569 from these southern populations had lower basal level of *Tnf-a* (a pro-inflammatory cytokine)
570 and higher level of *Mx2* antiviral gene expression than those from these northern populations
571 (Dubois *et al.*, 2018). Future studies should assess the potential relationships between
572 variations in gut bacteriota composition and the capacities to regulate or mount immune
573 responses and inflammation in these bank vole populations.

574

575 **Three-way relationships between intra-host communities**

576 We have not highlighted strong evidence of three-way relationships between the gut
577 microbiota (diversity or composition), the gastro-intestinal helminth and pathogenic bacteria
578 communities, when considering the whole communities. Neither the specific richness nor the
579 global composition of a given community are related to the richness or composition of the two
580 other ones. By contrast, particular taxa seem to be involved in these three-way relationships.
581 First, some infections are significantly associated with the global composition of the gut
582 bacteriota, but have only little impact on specific gut bacterial taxa. This result concerns two

583 helminths *Heligmosomum mixtum*, *Aonchotheca murissylvatici* and the hemotrophic bacteria
584 *Bartonella*. Opposite patterns are observed for the helminths and for the bacteria. They could
585 reflect the antagonistic impacts of these infections on the gut bacteriota, or negative
586 interactions between these pathogens. On one hand, some evidence suggests that *Bartonella*
587 may be acting as a symbiont more than a pathogen (refs in Lei & Olival, 2014). Significant
588 coevolutionary congruence has been found between *Bartonella* species and their rodent hosts,
589 and *Bartonella* infections in rodents lead to an asymptomatic long lasting intra-erythrocytic
590 bacteraemia (Deng *et al.*, 2012; Lei & Olival, 2014). It would be interesting to test whether
591 associations between *Bartonella* and gut bacteriota could corroborate the hypothesis of
592 coadaptation between these bacteria and their rodent hosts (Hayman *et al.*, 2013). On the other
593 hand, some hookworms have been shown to induce changes in rodent gut bacteriota (review
594 in Mutapi, 2015). Infection of mice with the nematode *Heligmosomoides polygyrus*, which is
595 phylogenetically close from *Heligmosomum mixtum*, lead to an increased abundance of
596 Lactobacillaceae in the gut microbiome (Reynolds *et al.*, 2014), as observed in our study.
597 Lastly, negative interactions between *H. mixtum* or *A. murissylvatici* and *Bartonella* are
598 probable, as gastro-intestinal hookworms are known to induce anaemia (Seguel & Gottsdenker,
599 2017), while *Bartonella* invades and replicates in red blood cells. This resource limitation
600 driven by helminths on erythrocyte-dependent infectious agents is an important driver of
601 helminth-microparasite coinfection (Graham, 2008). Therefore, the negative associations
602 detected here between *Bartonella* and *H. mixtum* or *A. murissylvatici*, and their respective
603 links with gut bacteriota composition, seem to be driven by potential complex antagonistic,
604 synergistic and symbiotic interactions that need to be further explored.
605 Second, other infections are strongly associated with large changes in the relative abundance
606 of one or few specific taxa from the gut bacteriota, but not with the global composition of this
607 later. These species-specific associations concern *Trichuris muris*, *Caetenotaenia henttoneni*,
608 *Paranoplocephala omphalodes* and *Mastophorus muris* for the helminths and *Neoehrlichia*
609 *mikurensis*, *Orientia tsutsugamushi*, *Mycoplasma haemomuris*, *Anaplasma phagocytophilum*
610 and *Rickettsia* for the bacteria.
611 Three bacterial infections (*Neoehrlichia* sp, *Orientia* sp and *Rickettsia* sp) as well as *P.*
612 *omphalodes* infections exhibited the same pattern: they were associated with a lower relative
613 abundance of Erysipelotrichaceae. It is striking to find such common associations for these
614 infectious agents because observed changes in the gut microbiota during infection are rarely
615 consistent, even with respect to single pathogens (Sabey *et al.*, 2021). The most obvious

616 features shared between these infectious agents is that they are transmitted by arthropods.
617 Unfortunately, there is still insufficient knowledge on Erysipelotrichaceae and its links with
618 infection or dysbiosis to explain the pattern observed. To our knowledge, such associations
619 have been investigated in humans only. They have shown that increased abundance of
620 Erysipelotrichaceae could be associated with a number of diseases such as tuberculosis, HIV
621 and norovirus infections, inflammation-related intestinal disease and metabolic disorders
622 (Kaakoush, 2015). The reasons why *Neoehrlichia* sp, *Orientia* sp, *Rickettsia* sp or *P.*
623 *omphalodes* infections are associated with a decreased abundance of Erysipelotrichaceae in
624 bank voles remain to be investigated.

625

626 **Marked relationships between gut bacteriota and gastro-intestinal helminth
627 communities**

628 While we do not detect three-way relationships between the whole composition of gut
629 bacteriota, helminth and pathogenic bacteria communities, we highlight strong pairwise
630 associations between helminth community and gut bacteriota. The former pattern might be
631 explained by the fact that the pathogenic bacteria detected here do not constitute a functional
632 community, in particular because their ecological niche can be very different. For example,
633 hemotrophic *Mycoplasma* parasitizes erythrocytes (Alabi *et al.*, 2020) while *Borrelia*
634 disseminates through the bloodstream and/or lymphatic system to invade and colonize various
635 tissue (Zeidner *et al.*, 2001).

636 The strong associations between gastro-intestinal helminths and gut bacteriota may be
637 interpreted under two perspectives. First, the strong positive associations between the
638 diversity of helminth community and gut bacteriota might corroborate the hypothesis and
639 experimental evidence showing that helminths have the capacity to maintain higher gut
640 microbiota diversity and may represent gut homoeostasis (Kreisinger *et al.*, 2015). Indeed,
641 low-intensity, chronic helminth infections are commonly linked to high microbial diversity
642 and predominance of bacteria typically associated with gut health (Peachey *et al.*, 2017).
643 Nevertheless, this interpretation has to be taken cautiously as the diversity of both
644 communities was strongly influenced by the localities of sampling. The environment might
645 therefore shape similarly gut bacteriota and helminth community diversity.
646 Second significant associations between helminth community and gut bacteriota composition
647 - which remain significant even when potential geographic confounding effects were removed
648 - may be linked to the fact that both communities reside in the same environmental niche

649 (host intestines). From there, they likely experience similar selection pressure (e.g., host
650 immune responses), with potentially strong interactions and reciprocal influence expected
651 between them, which could shape their composition (Glendinning *et al.*, 2014).

652 Unfortunately, the causal processes behind these gut microbiota and helminths interactions
653 are complex, multifaceted and difficult to assess, in particular because experimental studies
654 can only focus on single helminth infections— while interactions between/within community
655 are the rule within the host organism. The field of microbiota research would thus benefit
656 from taking into account the whole composition of gastro intestinal helminth community
657 rather than single helminth infections only.

658 In this study, we also highlight a large number of species-specific associations between
659 helminths infections and members of the gut bacteriota. High-intensity, acute helminth
660 infections may correlate with changes in hosts gut microbiota, through direct and indirect
661 (immune or other processes such as malnutrition) interactions (Peachey *et al.*, 2017).
662 Nevertheless, up to now, the patterns of shifts in gut bacteriota associated with helminth
663 infections remain hardly predictable. Research that addressed this issue, using laboratory or
664 wild animals, have provided variable and sometimes even contradictory conclusions, even for
665 single host-helminth models. A potential explanation is that these infection-associated
666 microbiota shifts could depend on the presence of other helminths and the duration of
667 infection {Sabey, 2021 #86 ; Schmid, 2022 #112}. Local interactions between helminths and
668 between helminths and gut bacteria could mediate changes in infection outcomes as well as
669 the gut bacteria and helminth populations themselves (Glendinning *et al.*, 2014).

670

671 Conclusion

672 Altogether, these results emphasize complex interlinkages between gut bacteriota, gastro-
673 intestinal helminths and bacterial infections in wild animal populations. We emphasize the
674 strong impact of environment, even at fine geographical scales, on these interactions. Shifts in
675 diet or host genetics could mediate the spatial changes observed in gut bacteriota. However,
676 the processes shaping gut bacteriota diversity and composition are many and complex, and
677 further investigations are required to decipher the relative importance of drift, dispersal or
678 selection on bank vole gut bacteriota in the populations studied here. Besides, we find a
679 diverse array of associations between gut bacteriota and gastro-intestinal helminths or
680 pathogenic bacteria, some being significant at the scale of the whole community and other
681 being species-specific only. Whether these patterns reflect coadaptation, dysbiosis or indirect

682 interactions with host immunity and coinfections should now be considered to better
683 understand the spatial variations observed in the relationships between gut bacteriota and
684 health.

685

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690

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696

697 **Conflict of interest disclosure**

698 The authors declare that they have no financial conflict of interest with the content of this
699 article. N. C. is one of the PCI Inf recommenders. B.R. is part of the managing board of PCI
700 Inf.

701

702 **Data and script availability**

703 Raw data and scripts are available on zenodo: <https://doi.org/10.5281/zenodo.7433800>

704 705 **Supplementary information**

706 All supplementary materials are available on zenodo :

707 <https://doi.org/10.5281/zenodo.7431842>

708

709 **Supplementary Figure S1.** Maps showing the sampling area (left) and localities (right) in
710 France. Forests are indicated in green and water in blue. The four sampling localities are
711 represented with a colored polygon. The arrow indicates the North.

712

713 **Supplementary Figure S2.** Composition of the gut bacteriota. The relative abundance of six
714 phyla representing 99% of the total composition is represented. Individuals are grouped by

715 sampling localities, which are ordered from North to South. (A) Bar graph shows individual
716 variation in phyla composition (phylum=color). (B) Box and whisker plots represent median
717 and interquartile values for each phylum. Black dots correspond to mean values, and colored
718 dots correspond to individuals.

719
720 **Supplementary Figure S3.** Variations of alpha diversity with individual factors, for the gut
721 bacteriota (family level), pathogenic bacteria and gastro-intestinal helminths of bank voles.
722 Alpha diversity is estimated using the specific richness (A, B and C) and the Shannon index
723 (D, E and F). In graphs C and F, the blue line corresponds to the linear regression line.

724
725 **Supplementary Figure S4.** Relationships between the composition of the gut bacteriota,
726 pathogenic bacteria and gastro-intestinal helminth communities: The db-RDA triplot shows
727 the structure of the gut bacteriota at the phylum level and the correlations with the intra-host
728 parasite communities. The arrows correspond to the significant explanatory variables. Each
729 point corresponds to an individual, and the colors correspond to the different sampling
730 localities.

731
732 **Supplementary Table S1.** Variation of the Firmicutes/Bacteroidetes ratio with localities and
733 individual factors.

734
735 **Supplementary Table S2.** Alpha diversity metrics and statistics for the gut bacteriota,
736 pathogenic bacteria and helminth communities of bank voles.

737
738 **Supplementary Table S3.** Beta diversity metrics and statistics for the gut bacteriota,
739 pathogenic bacteria and helminth communities of bank voles.

740
741 **AUTHOR CONTRIBUTIONS**

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